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Title: An Exploration of Ensemble GI software using LANL TA-66 sensor data

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# **An Exploration of Ensemble** GI software using LANL TA-66 sensor data

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### **Outline**

- Process of testing
- Smoothing Data
- Data gap (missing data) experiment, no resampling for sparsities
- Data gap (missing data) experiment, resampled data for dense time
- Conclusion

- Note about graph labels in the following slides:
  - x axis is always time
  - y axis is sensor value, except on discrete quantile bin plots, which are ordinal, i.e., the nth varying width quantile bin, not percentages



## **Process of testing**

- Acquired Ensemble Grammar Induction (GI) software
- Tested with raw data using two years of time series
- Consulted with Constantin Brif.
- Smoothed data with Lasso
- Tested 3 Lasso'd sensor data sets in Ensemble GI GUI
- Data gap (missing data) experiment
  - TA-66 sensor data has missing values
  - Find data gaps as anomalies? sparse time
  - Find data gaps as anomalies? dense time (reinterpolating) the data in time

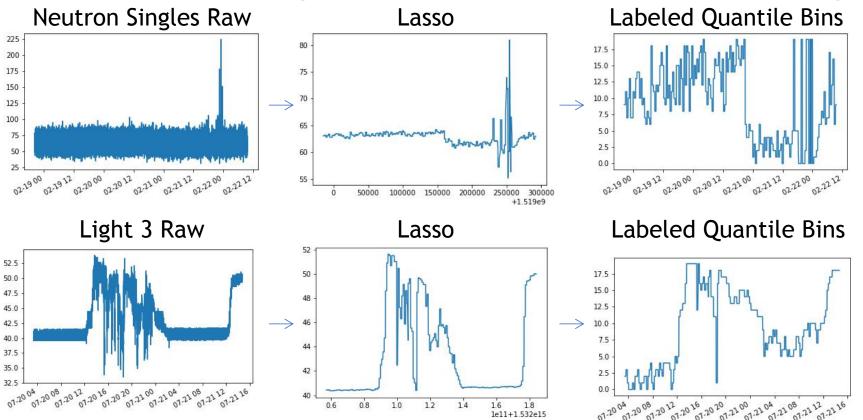


## Smoothing data, i.e., reducing the data

- Lasso (Least Absolute Shrinkage and Selection Operator) + Lars (Least Angle Regression)
  - Piecewise linear regression; used before on similar types of MINOS sensor data
  - Lines can be sloped, depending on Lasso parameter tuning
- Manual symbol (label) generation from Lasso line segments
  - Run Lasso to generate at most 1 line segment per 1 minute: there may be fewer segments where one line spans multiple minutes, depending on goodness of fit
  - Generate one feature per minute as the average of the line over that minute
  - Preselect a number of symbols (n) and discretize each minute into n varying width quantile bins to make sensor values uniform (e.g., k bin discretizer)
- Ensemble GI settings to match
  - Piecewise aggregate approximation (PAA) = 1, no need to further linearize
  - Number of symbols =  $\mathbf{n}$ , no need to further discretize



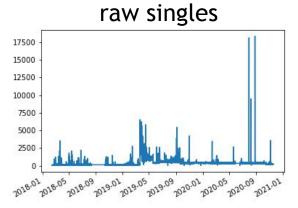
# 1 minute Lasso segments and 20 bins over several days

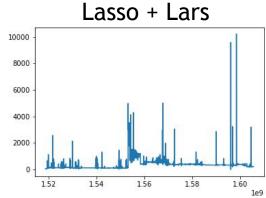


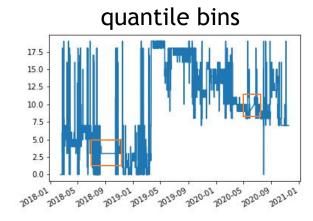


### **Data Set 1**

- Neutron detector
  - 25 million rows, two columns
  - 2 years at ~3 second intervals, single and double counts; example gaps in data
  - Smoothed to 1 minute intervals (20:1 in time), 20 discrete quantiles



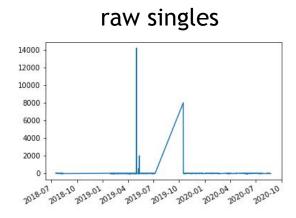


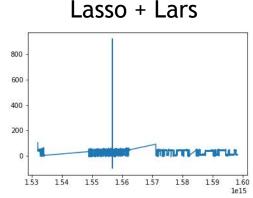


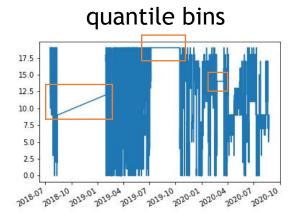


### Data Set 2

- Light Sensor 3 (light closest to neutron detector)
  - Raw 31 million rows, one column
  - 2 years of ~1 second intervals, light intensity; example gap(s) in data
  - Smoothed to 1 minute intervals (60:1 in time), 20 discrete quantiles





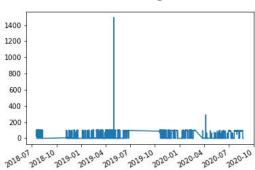




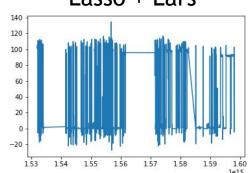
### **Data Set 3**

- Light Sensor 6 (1 of 3 conference room sensors)
  - 35 million rows, one column
  - 2 years of 1 second intervals, light intensity; example gap(s) in data
  - Smoothed to 1 minute intervals (60:1 in time), 9 discrete quantiles

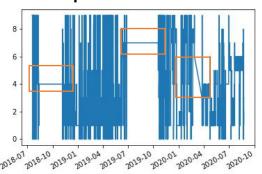








### quantile bins





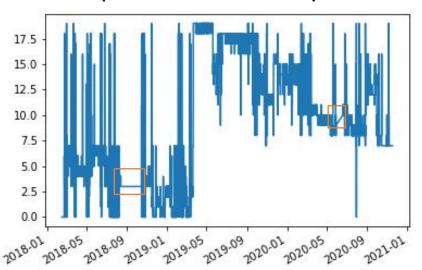
## Experiment to find missing data, irregular time

- All of the TA-66 data streams do not have regularly sampled data
  - Data are sparse, in general, due to possible lag in sensor timing or events
  - Also, power outage, sensor/platform failure, data collection interrupted, etc.
  - Our data are time stamped for each sample: i.e., you cannot infer real time from sample position in the stream due to irregular sampling
- Curious to see if the anomaly detector could flag missing data
  - Tested sensor data as is with Ensemble GI: e.g., the streams were unevenly sampled in time, due to drops or just sensor time lag between samples
- Experiment 1: Would missing data be flagged without real time?
  - Would discontiuities in the value space (sensor readings) "a data drop blip" would be noticable without the time index/being densely, regularly sampled?

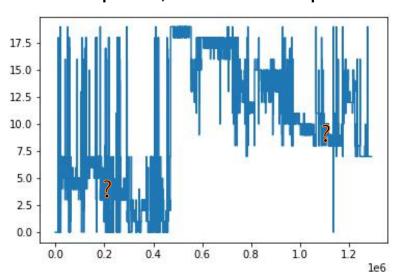


# "Compressed in time": i.e., sparse real time vs. indexed time

sparse but timestamped



### sparse, no timestamp



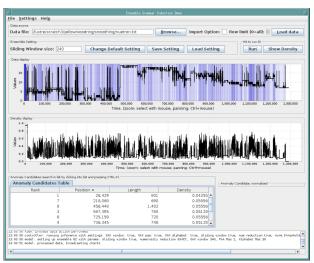


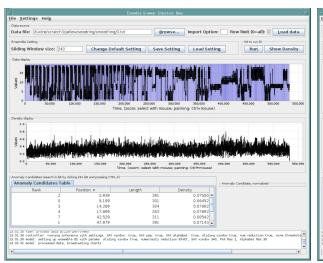
# Running Ensemble GI with sparse time data

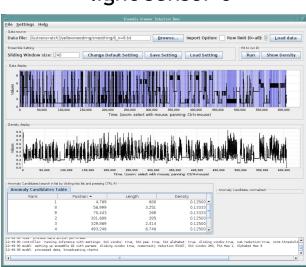
neutron singles

light sensor 3

light sensor 6







- 3 trials, 10 top anomalies, windows of 2 minutes, 15 minutes, 4 hours, 1 day
- Didn't seem to find anomalies to that matched time of gaps (User error?)

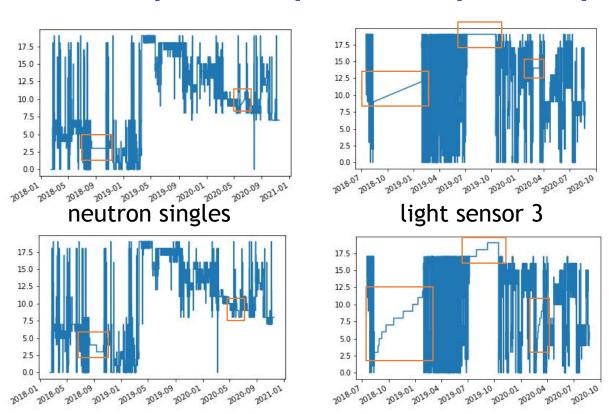


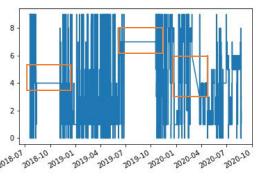
## Rerun experiment with dense sampled in time data

- User error? Should have just smoothed and downsampled without uniform binning? Unanswered questions
- Experiment 2: Rerun, but with regular/evenly spaced samples in time
  - Same settings for Lasso and quantile binning
  - Same settings for Ensemble GI, including number of trials and top anomalies
  - Additional data for Ensemble GI to process, but no noticable addition in time
- Resample the data to have regular (even) sampling in time
  - Neutron data, reinterpolated to have exactly 1 sample / 3 seconds
  - Light sensor 3 and light sensor 6, interpolated to have exactly 1 sample / 1 second
  - Filled in gaps with samples, but those samples were anomalous wrt other data

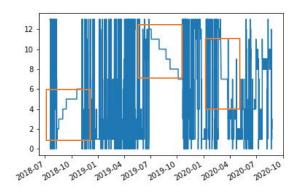


## Resampled comparison - sparse top, dense bottom





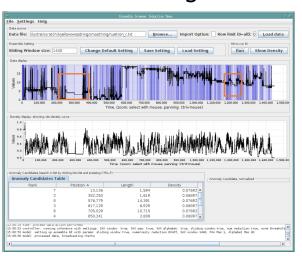
light sensor 6





# Running Ensemble GI with dense time (regular Hz)

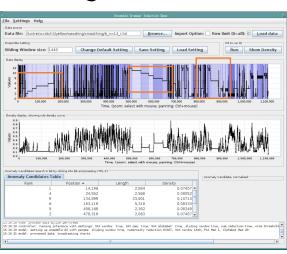
neutron singles



light sensor 3



light sensor 6



- 3 trials, 10 top anomalies, windows of 2 minutes, 15 minutes, 4 hours, 1 day
- With a time window of 1 day: Ensemble GI flagged the missing data, i.e., low rule density, i.e., anomalies, correspond to the missing data



# Example timings of missing data compared to anomalies

### Missing data (from, to)

- Neutron detector
  - 2018-08-13 12:10:43 to 2018-10-12 21:35:16
- Light sensor 3
  - 2018-08-12 23:59:59 to 2019-01-29 06:12:56

### **Anomalies (initial, length in minutes)**

- Neutron detector
  - **-** 2018-08-14 12:11:00 41599

- Light sensor 3
  - **-** 2018-08-14 01:07:55 16270



### Conclusion

- Infrastructure in place to test with TA-66 data and first high-level results
- Fast anomaly detection with 2 years of data at 1 minute (< 30 seconds)</li>
- Future Work
  - More testing
  - Integrate Ensemble GI to generate features
  - Build results into feature matrix
  - What to do about sparse data, in general?
  - Correlate anomalies across sensors and time scales

